

MBC 638

DATA ANALYSIS AND DECISION MAKING

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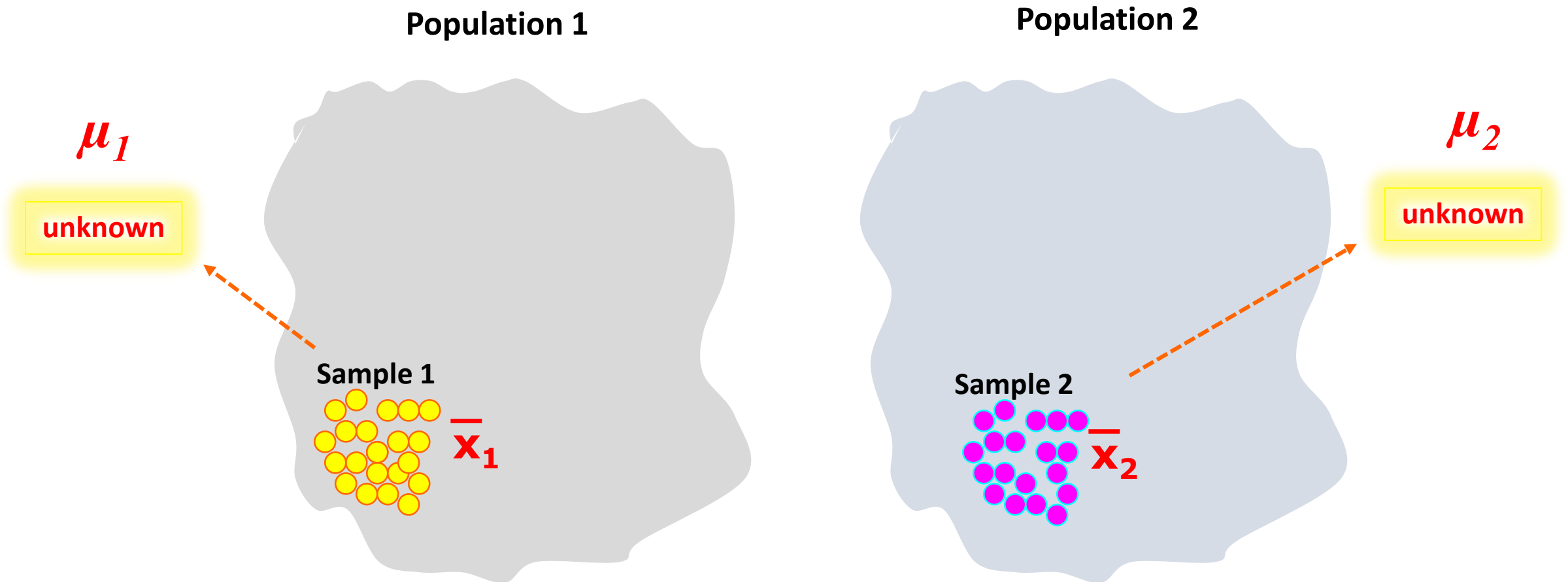
HYPOTHESIS TESTING for the difference in population means

Chapter 9

Today:

Hypothesis testing for $\mu_1 - \mu_2$

- We use information from 2 observed samples to conclude how μ_1 and μ_2 compare.
- Assumption: The two samples are **independent**. (We will relax this assumption later.)



Example

We want to test hypothesis that, on average, data scientists that have a graduate degree μ_1
earn a higher salary than those that don't have a graduate degree. μ_2

$$H_0: \mu_1 \leq \mu_2$$

$$H_A: \mu_1 > \mu_2$$

It's the same as:

$$H_0: \mu_1 - \mu_2 \leq 0$$

It's the same as:

$$H_A: \mu_1 - \mu_2 > 0$$

Hypothesized
mean
difference
under the null
hypothesis

Recall: In hypothesis testing for μ

Test statistic is the **Z-score** of the observed **sample mean**

$$\frac{\bar{x} - \mu_0}{\sigma / \sqrt{n}}$$

Analogously, in hypothesis testing for $\mu_1 - \mu_2$

Test statistic is the **Z-score** of the **difference** between the observed **sample means**

$$\frac{\bar{x}_1 - \bar{x}_2 - 0}{\sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}}$$

Hypothesized mean difference under the null hypothesis

| | | |
|-----------------------|--------------------------|-------------------------------|
| Left-tail test | $H_0 : \mu_1 \geq \mu_2$ | $(i.e. \mu_1 - \mu_2 \geq 0)$ |
| | $H_A : \mu_1 < \mu_2$ | $(i.e. \mu_1 - \mu_2 < 0)$ |

| | | |
|------------------------|--------------------------|-------------------------------|
| Right-tail test | $H_0 : \mu_1 \leq \mu_2$ | $(i.e. \mu_1 - \mu_2 \leq 0)$ |
| | $H_A : \mu_1 > \mu_2$ | $(i.e. \mu_1 - \mu_2 > 0)$ |

| | | |
|--------------------|--------------------------|-------------------------------|
| 2-tail test | $H_0 : \mu_1 = \mu_2$ | $(i.e. \mu_1 - \mu_2 = 0)$ |
| | $H_A : \mu_1 \neq \mu_2$ | $(i.e. \mu_1 - \mu_2 \neq 0)$ |

$$p\text{-value} = P\left[z < \frac{\bar{x}_1 - \bar{x}_2}{\sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}} \right]$$

$$p\text{-value} = P\left[z > \frac{\bar{x}_1 - \bar{x}_2}{\sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}} \right]$$

$$p\text{-value} = 2 \bullet P\left[z > \left| \frac{\bar{x}_1 - \bar{x}_2}{\sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}} \right| \right]$$

If σ_1 and σ_2 are unknown, use **t**-distribution (or Z approximation if n_1 and n_2 are both ≥ 30)

Example

Periodically, Merrill Lynch customers are asked to evaluate Merrill Lynch financial consultants and services (*2000 Merrill Lynch Client Satisfaction Survey*). Higher ratings on the client satisfaction survey indicate better service, with 7 the maximum service rating.

Independent samples of service ratings for two financial consultants—Ms. Chen and Mr. Wang are summarized here. Ms. Chen has 10 years of experience, whereas Mr. Wang has 1 year of experience.

| | Ms. Chen (10 yrs) | Mr. Wang (1 yr) |
|---------------------|-------------------|-----------------|
| # reviews collected | 36 | 30 |
| Mean | 6.58 | 6.25 |
| Standard deviation | 0.64 | 0.75 |

- At $\alpha=5\%$, determine whether consultants with more experience get **higher** average service rating.
- At $\alpha=5\%$, determine whether consultants with the two different levels of experience get **different** average service ratings.

What if the samples are **not independent**?

Paired samples

Example

Sales Presentation Ratings.xlsx

Example

Task Times.xlsx

Today: Hypothesis testing for $\mu_1 - \mu_2$

■ Independent samples



■ Dependent (paired) samples

